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Regressions in AIRS v5 Retrieval

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This work was carried out at the Jet Propulsion Laboratory, California Institute of Technology under a contract with the National Aeronautics and Space Administration.



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Summary

- **Murty Divakarla of NOAA and Thomas Hearty of NASA have shown spurious trends ~ 100 mK/yr in version 4 & 5 AIRS retrievals vs. truth**
- **Evidence points to regression retrieval steps as a major source of these**
- **Version 6 AIRS retrievals will reduce reliance on regressions and improve practices where regressions are retained**



From Divakarla -- Apparent Trend in AIRS v4 vs. Radiosonde

- Divakarla et al 2006
- Correlated with CO₂
- AIRS version 4
- AIRS version 5 added changing CO₂ background in physical retrieval

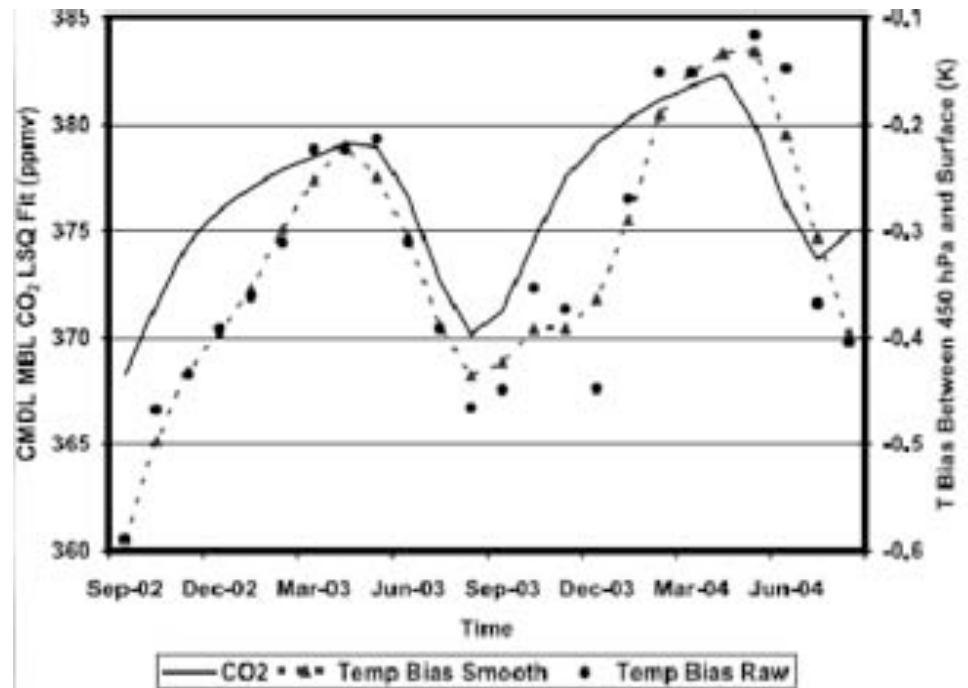
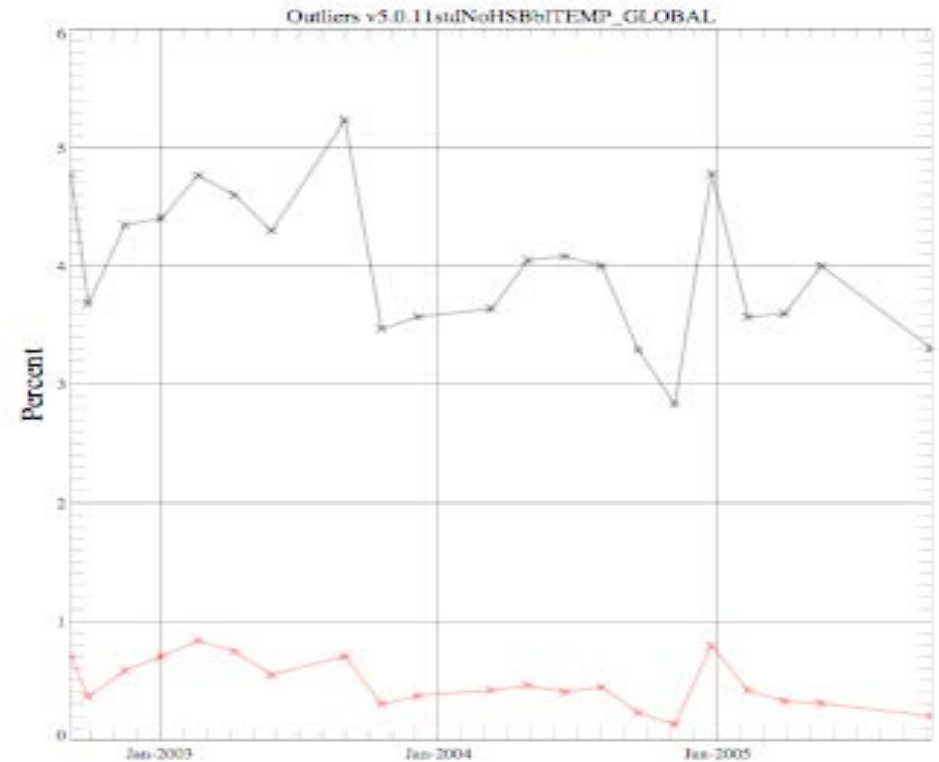
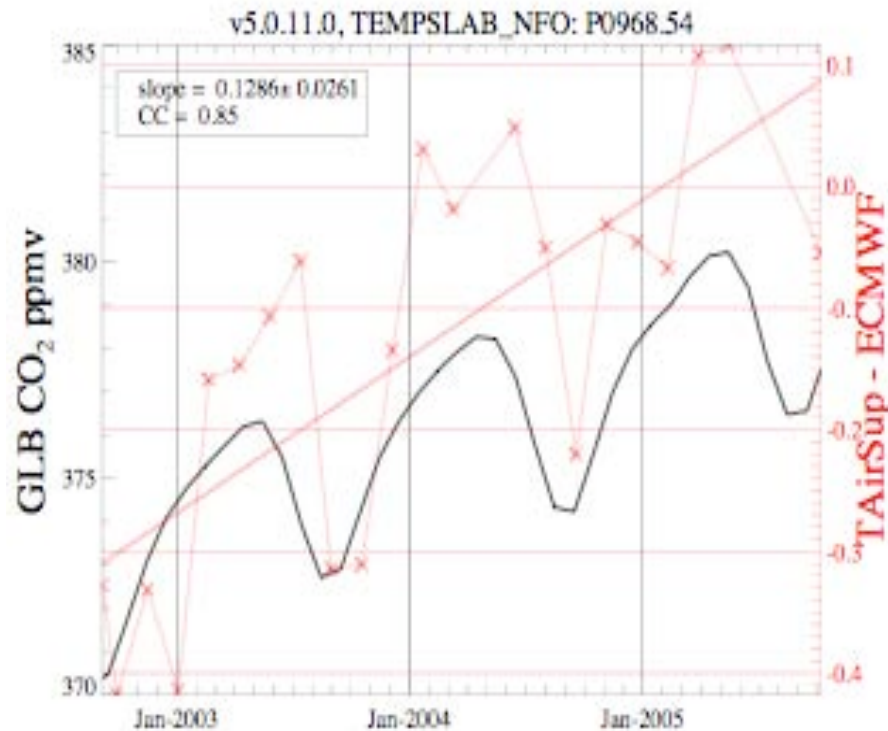


Figure 14. Seasonal trends between AIRS retrieval bias 150 hPa to surface and CMDL MBL CO₂, 90°N–90°S. Average differences between RAOB and AIRS temperatures are indicated by solid circles, smoothed differences using a 2-month sliding boxcar average are indicated by the dashed line, and zonally weighted linear least squares fit for the CMDL MBL product are indicated by the solid line.



From Hearty - Trend in V5 Global Temperature

- Upward trend in temperature bias vs. ECMWF
- Downward trend in outliers



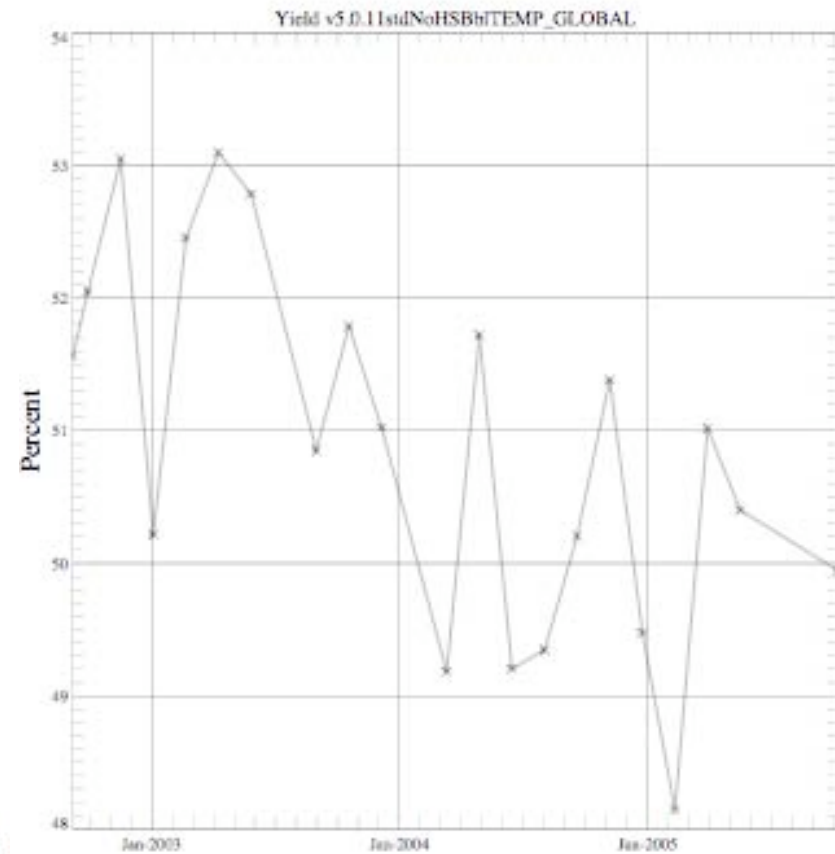
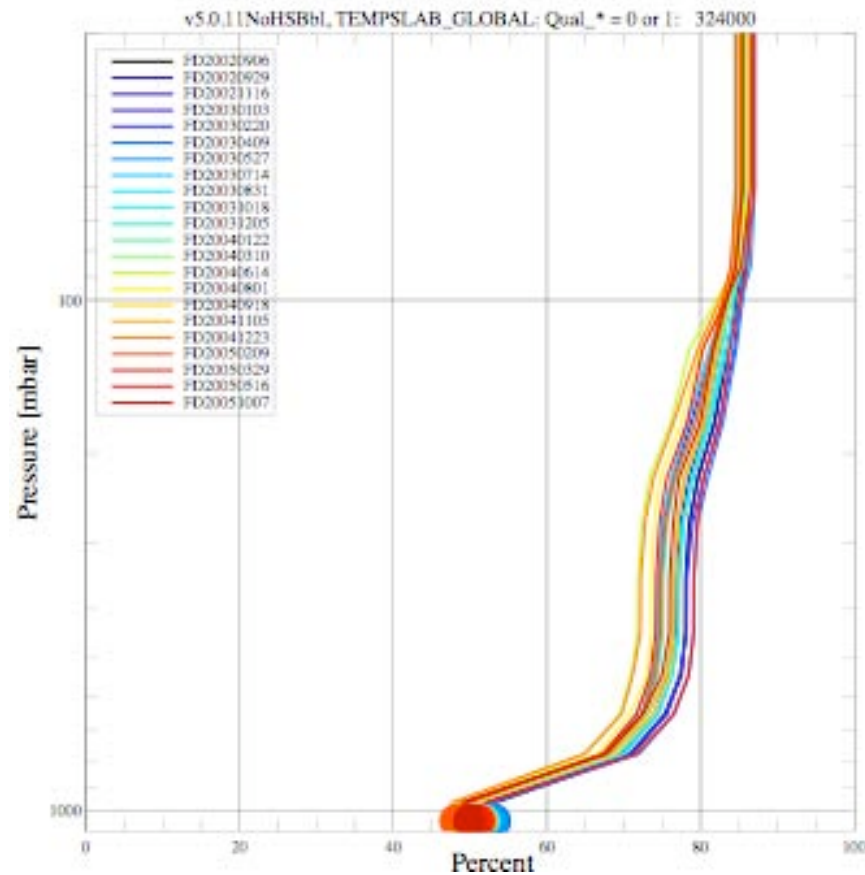
- Black line are mild outliers
- Red line are extreme outliers



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From Hearty - Trend in V5 Global Temperature Yield



Much more in Hearty presentation in <http://airs.jpl.nasa.gov/Science/ResearcherResources/MeetingArchives/TeamMeeting20070327/>



Temporal Variation in Local Angle Adjustment

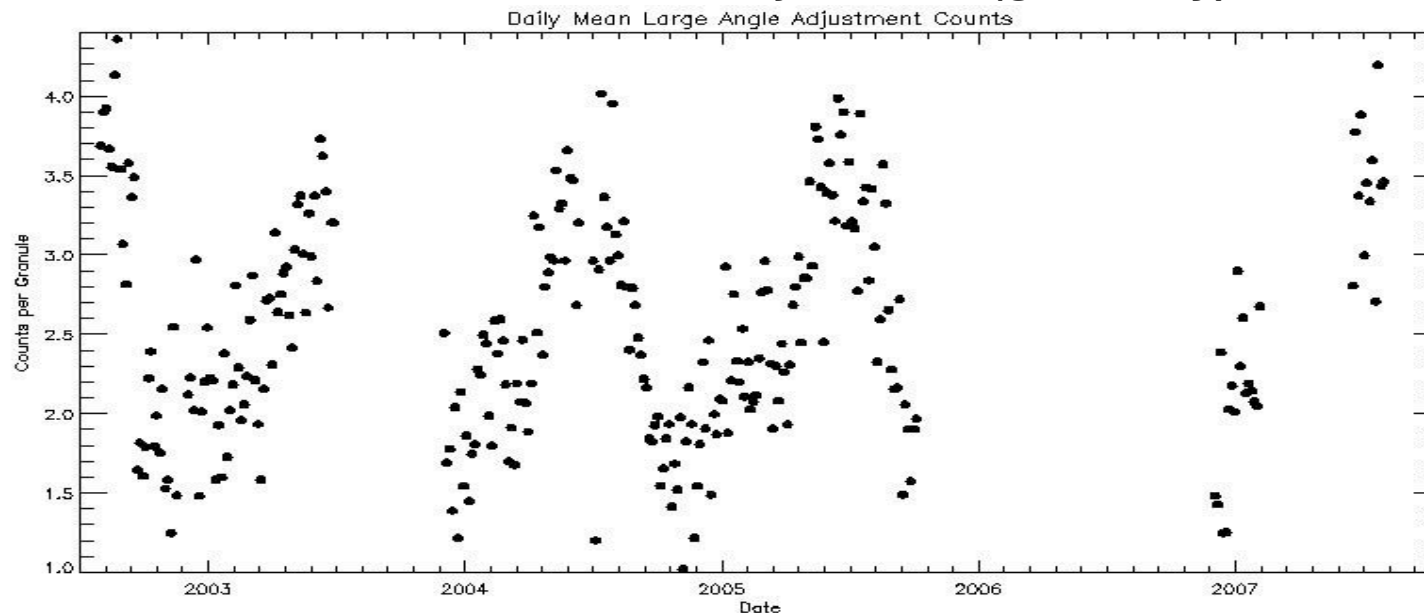
Background:

- **AIRS v5 retrievals are performed over a 3x3 array of FOVs, assuming all differences among the 9 FOVs are due to clouds**
- **Because of the instrument scan pattern, these 9 FOVs are observed at 3 different angles through the atmosphere, introducing small differences in the spectra**
- **Local angle adjustment makes small changes to the spectra from the outer 6 FOVs to emulate what would have been seen at the central angle**



Temporal Variation in Local Angle Adjustment

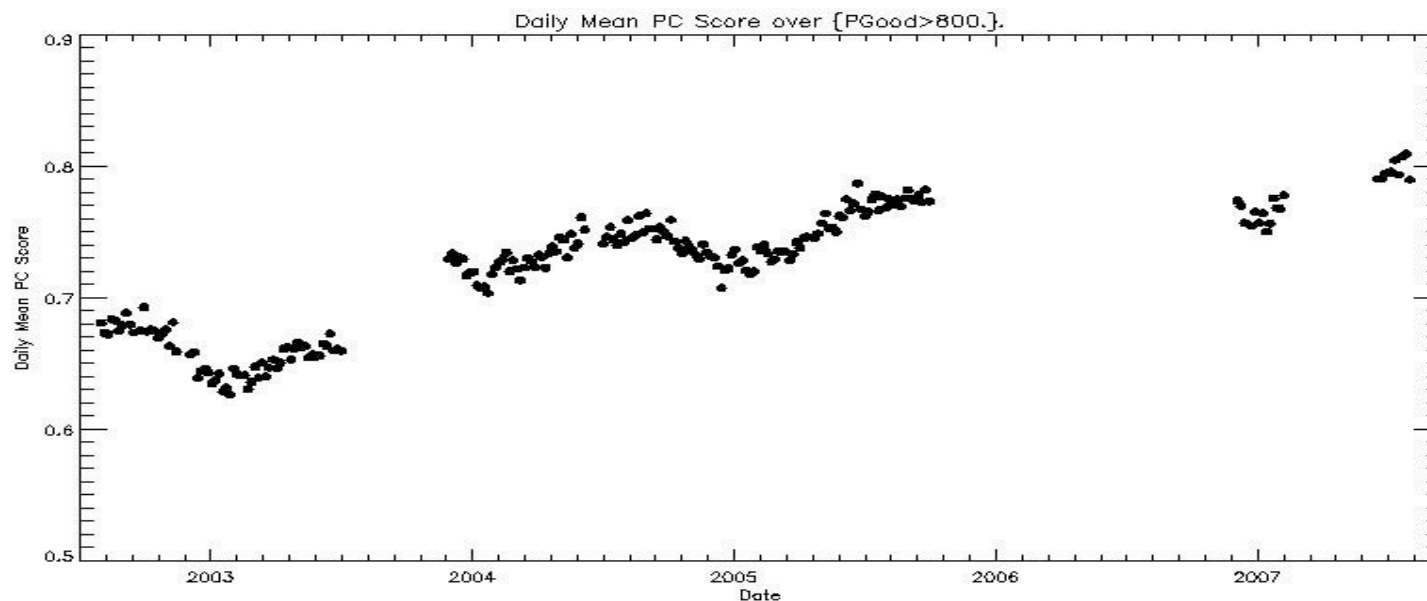
- Each 6-minute granule produces a count of number of FOVs with "big" angle adjustments (at least 5 channels adjusted by at least $20 \times \text{noise}$)
- The number of these cases shows a strong annual cycle
- But remember, LAA is a small adjustment (generally)





Temporal Variation in PC Scores

- **Lower PC score means the input matches the training set better**
- **PC Scores are rising with time**
- **There is a clear seasonal cycle**
- **PC Scores are used in quality control -- higher PC scores mean more rejections.**



Daily Mean of PC Scores where Pgood > 800 mbar



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Why Suspect Regressions?

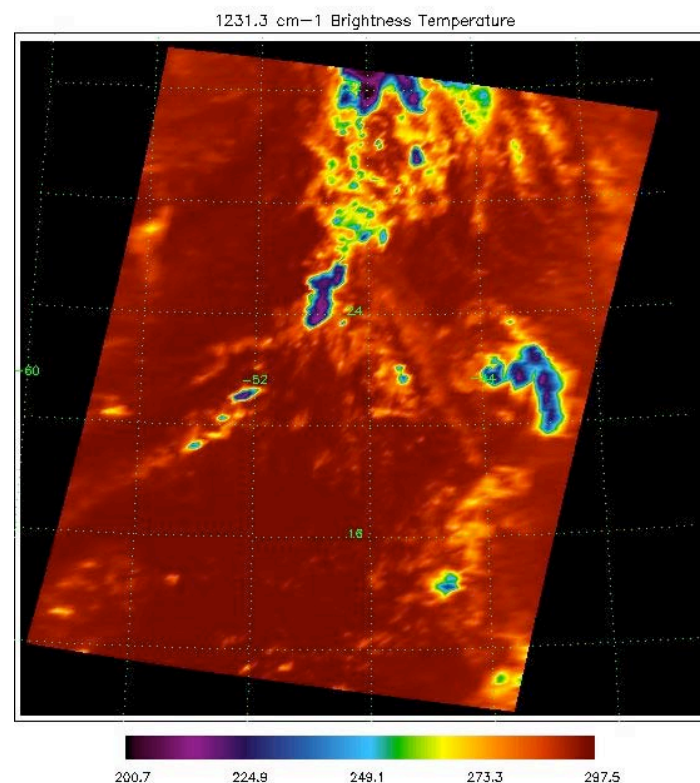
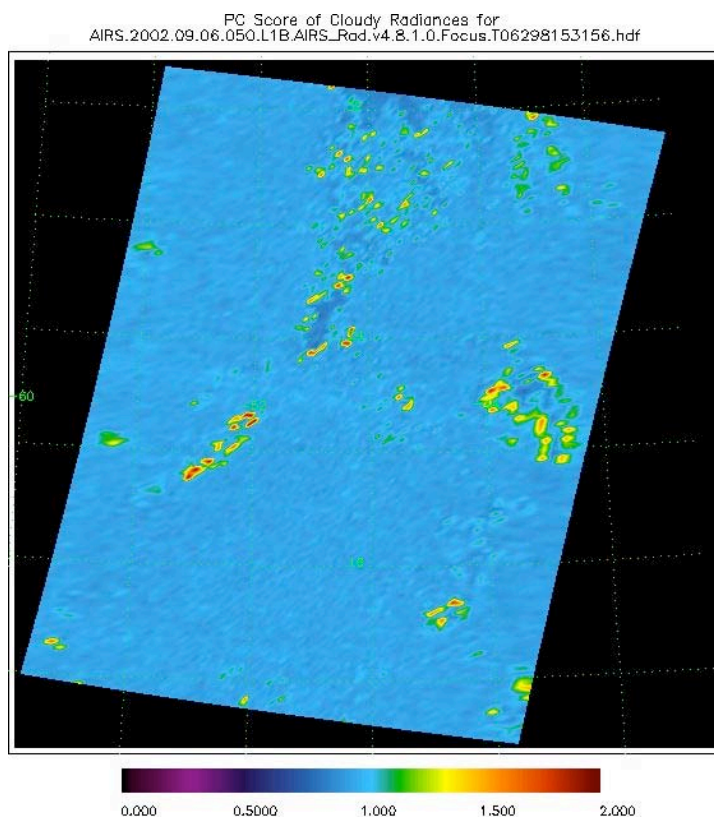
- **Regressions occupy key points in the retrieval**
- **Regressions have a known dependency on training data -- they only know how to handle what they have seen before**
 - These regressions use a large number (~50%) of all 2378 AIRS channels. When any channel is unavailable, it must be filled somehow.
 - PC Scores are consistently elevated in regions of fires, dust, edges of clouds, sun glint, SO₂, etc.
 - Regressions are trained with a narrow range of background CO₂ will have trouble with later data with more CO₂



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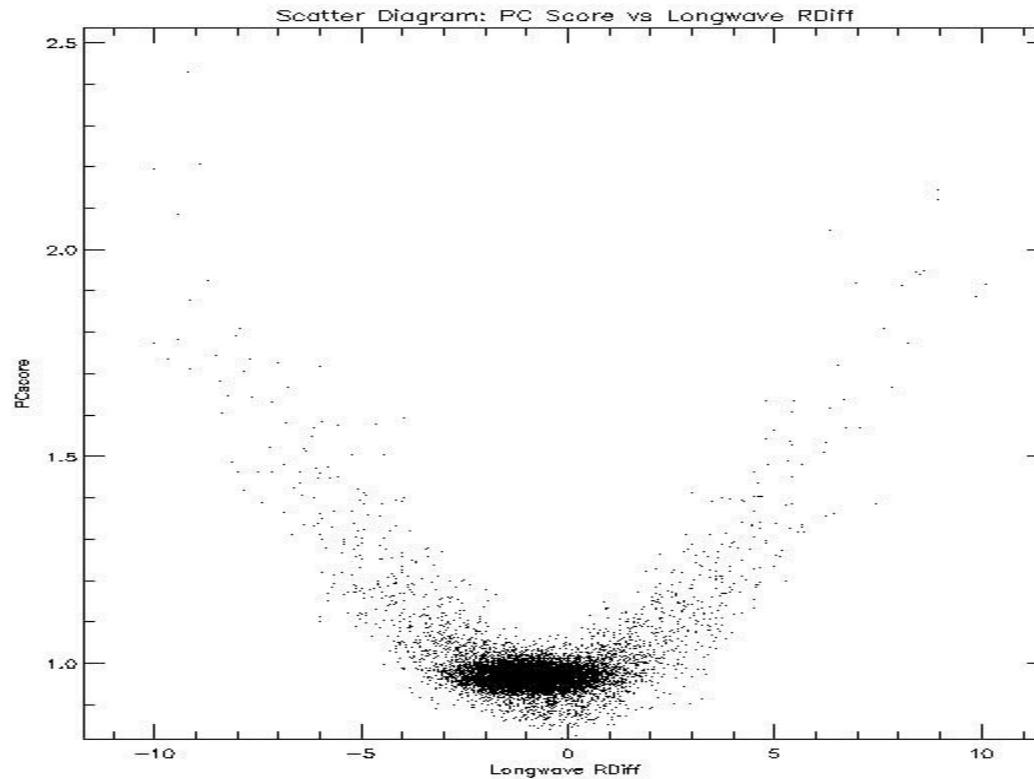
Difficult Cases for Regression -- Edges of Clouds



PC Score Granule 50 of Sept 6, 2002 Tb 1231 cm⁻¹
High value of PC score is correlated with side of cloud, where C_{ij}
tends to be high



Difficult Cases for Regression -- Edges of Clouds (Cont)



**Scatter diagram of PC Score vs. Longwave Rdiff, a
measure of C_{ij}**



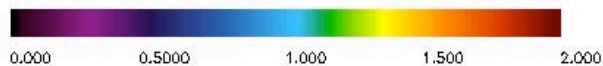
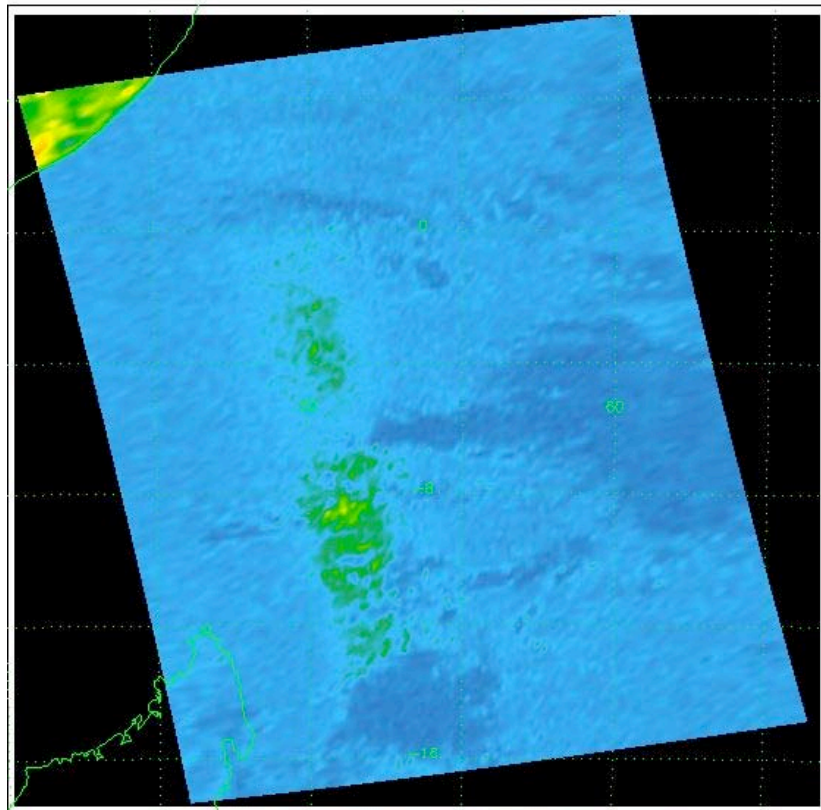
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Difficult Cases for Regression -- Sun Glint

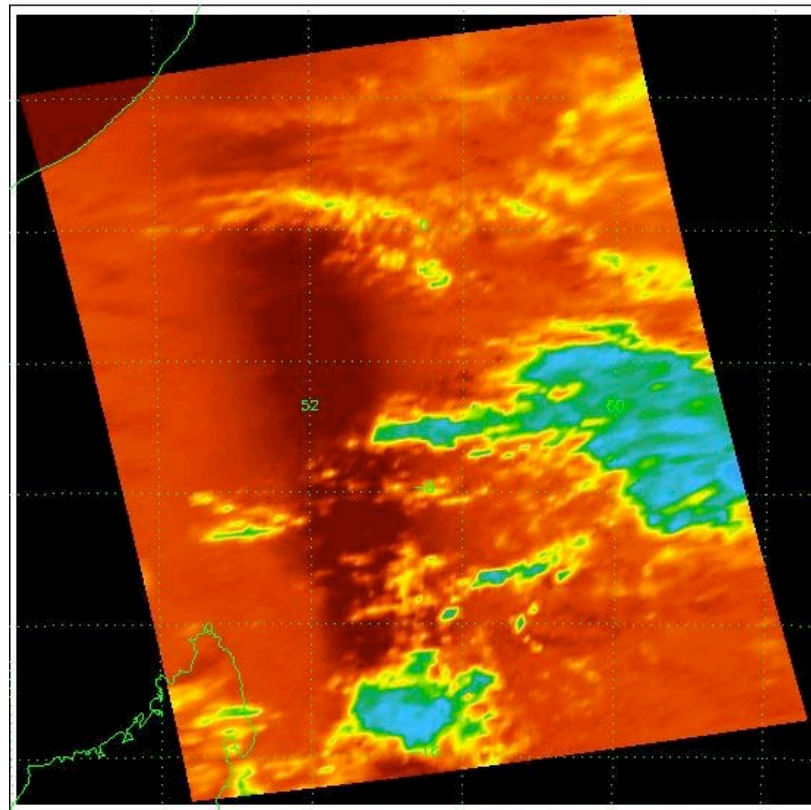
PC Score

PC Score of Cloudy Radiances for
AIRS.2003.03.02.099.L1B.AIRS_Rad.v5.0.0.0.G07079074048.hdf



2616 cm⁻¹ Br Temp

Brightness Temperature of 2616.4 cm⁻¹ Channel for
AIRS.2003.03.02.099.L1B.AIRS_Rad.v5.0.0.0.G07079074048.hdf

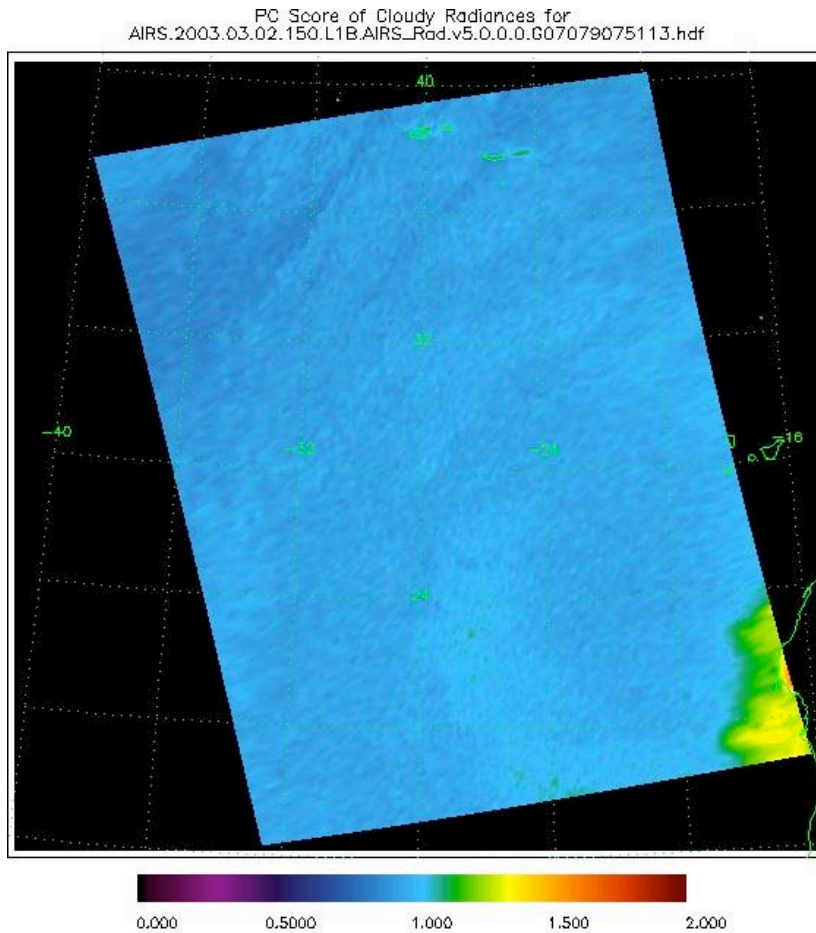


Granule 99 of March 2, 2003

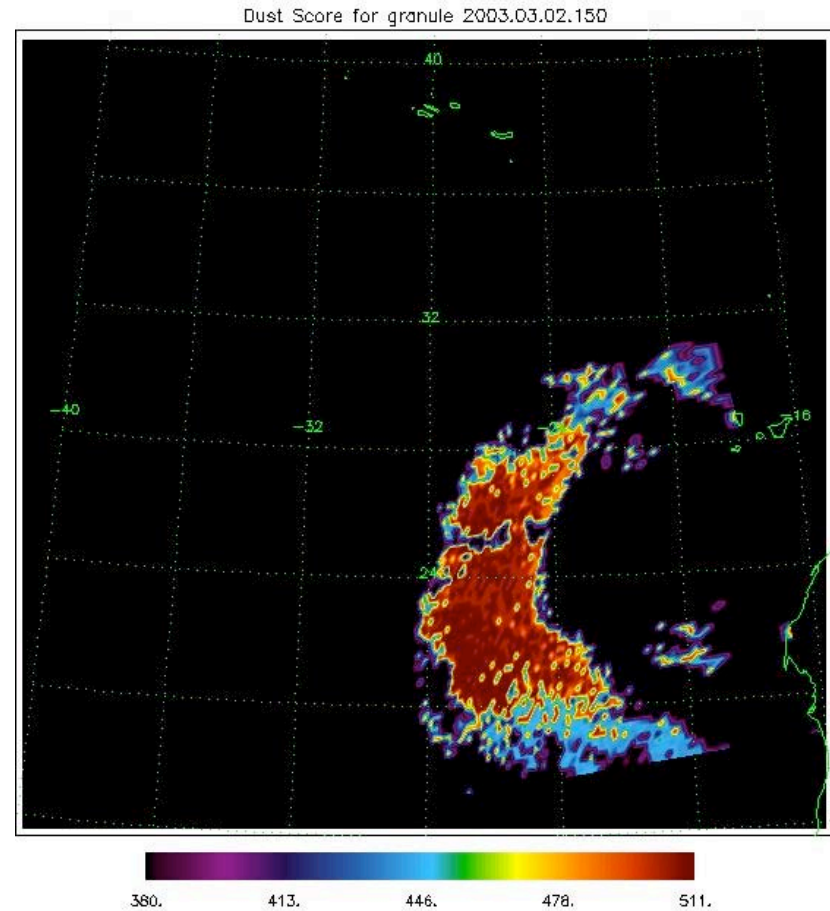


Difficult Cases for Regression -- Dust

PC Score



Dust Score

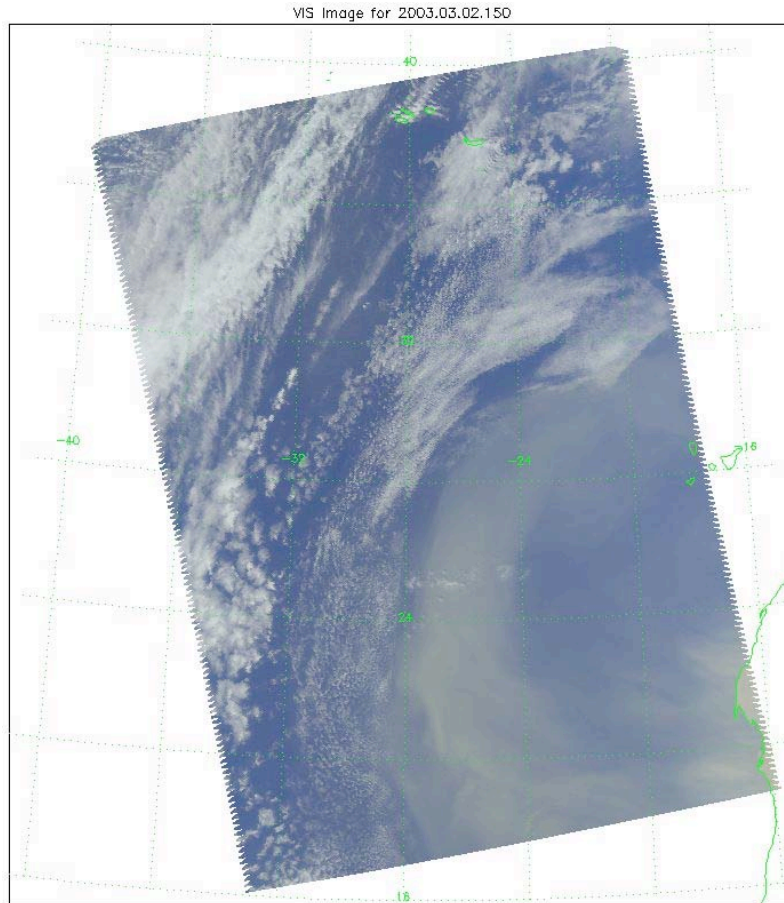


Granule 150 of March 2, 2003



Difficult Cases for Regression -- Dust (cont)

Dust Score Misses Some Dust



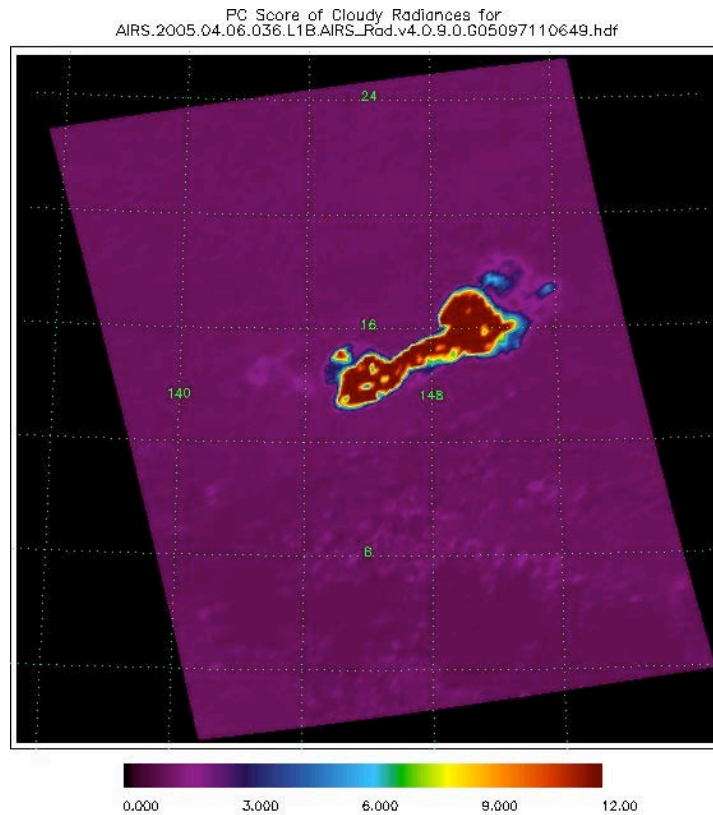
- **Dust plume near nadir is detected by dust score**
 - Only marginally high PC Score
- **Dust plume near the southeastern corner of granule is missed by dust score.**
 - Large PC Score



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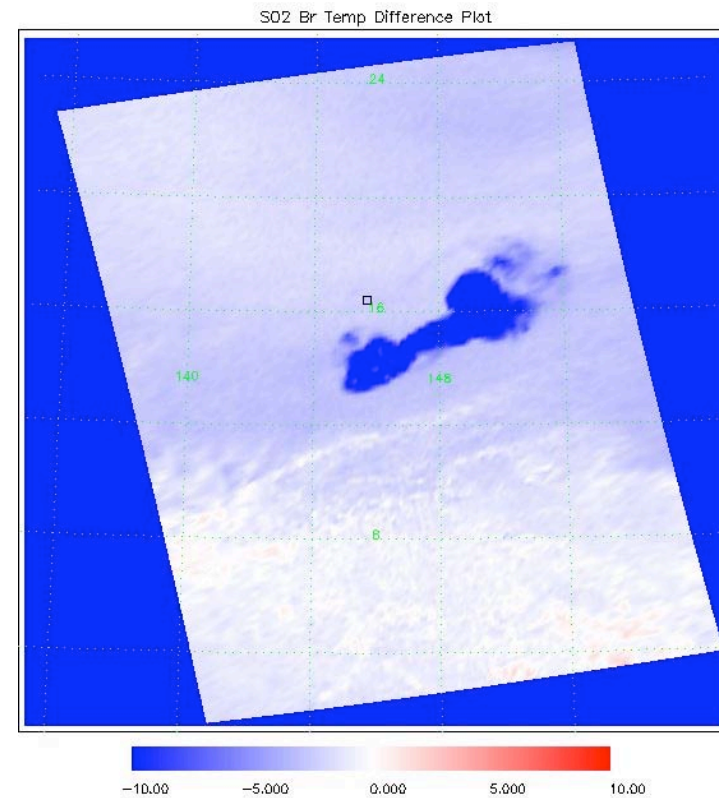
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Difficult cases for Regression -- SO₂



PC Score

- Volcanic plume from Anatahan
- Granule 36 of April 6, 2005

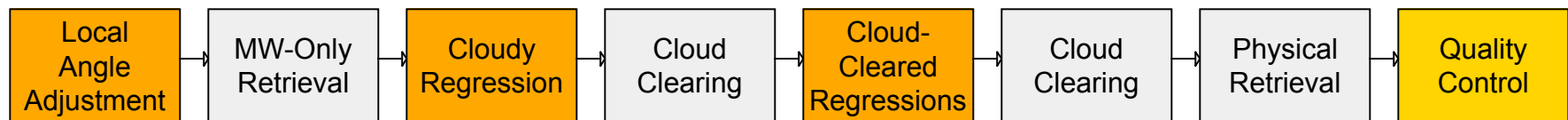


SO₂ Br Temp Diff



Placement of Regressions

- **AIRS retrieval includes these key regression steps:**
 - Local angle adjustment
 - 1st guess cloudy regression
 - Cloud-Cleared profile regression
 - Cloud-Cleared surface property regression
- **Cloud Clearing plus physical retrieval as last retrieval step should attenuate the impact of upstream regressions**
- **Quality control mixes in regression results**
 - Uses PC scores
 - Uses differences between results of regressions and physical retrieval





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Channel Filling

- **Radiances of channels needed by regression are replaced with synthetic radiances when those channels are not considered useable.**
- **Overzealous standards have led to too many channels being filled. This will be reduced in version 6.**
- **The current channel filling algorithms are not optimal. They will be updated in v6.**
- **See details in backup material.**



AIRS Channel Filling -- First 4+ Years

Year	Number of Channels Routinely Filled (out of 1680)
Late 2002-2003	1 - 7
2004	3 - 7
2005	2 - 14
2006-Early 2007	6 - 16

Spot check of 1st scan of granule #120 of selected focus days



Tests of Channel Filling

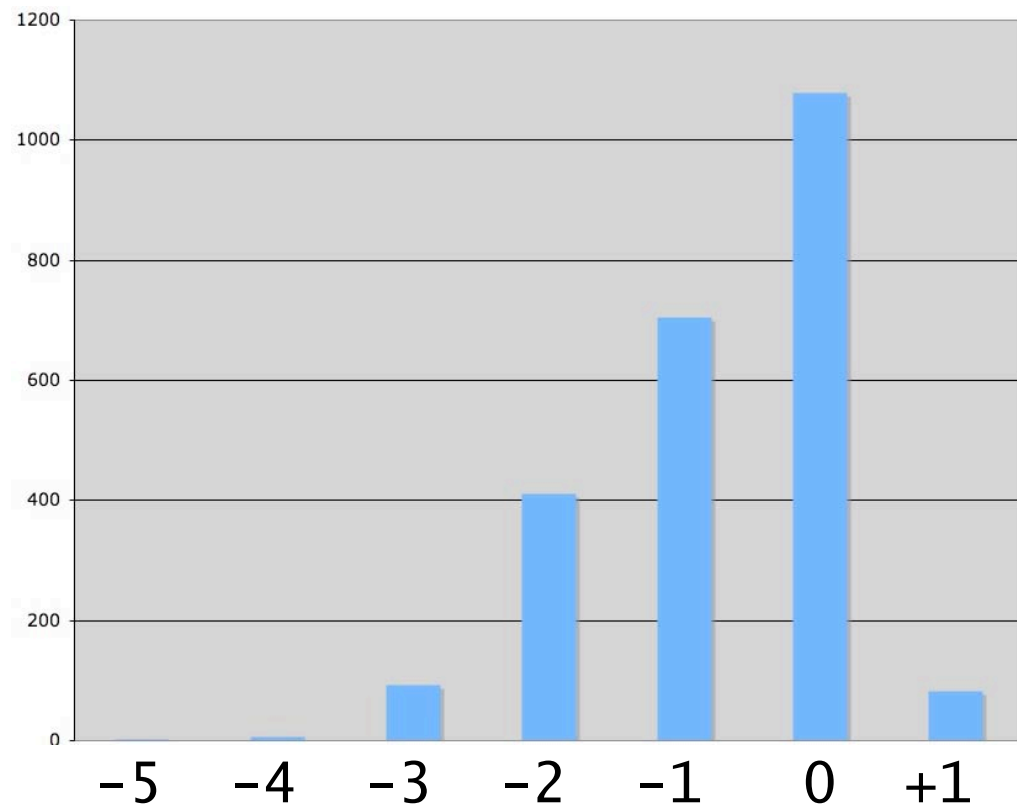
- **These tests selectively block channels in Level-1B radiances and look at results of full retrieval**
 - Test 1
 - One granule is run 2378 times, with one channel flagged bad each run
 - Test 2
 - Data for 2002-09-06 (focus day 3) was run twice and results were compared:
 - 1st run is exactly released v5.0 product
 - 2nd run uses the v5.0 algorithm but the input is changed -- 15 channels which are not used on 2005-01-30 are flagged bad in the Level-1 input to retrieval



Results of Channel Filling Test 1

- **Histogram of change in yield of retrieval-type 0 (out of ~1000)**

Filling a single
“average”
channel causes
yield to drop by
~0.1%





Results of Channel Filling Test 1

Channels with the largest effect on total cloudiness --
bias in *mean* cloudiness over an entire granule

Chan #	Freq cm ⁻¹	CC 1	CC 2
2109	2388.2	17 %	8 %
2110	2389.1	21 %	7 %
1876	2182.3	-7 %	2 %
1871	2186.9	-10 %	2 %

- **The worst channels to lose are those near gaps in the regression set.**
 - Planned changes to the filling algorithms will fix this.
- **Physical retrieval reduces but does not eliminate the effect.**
- **Fortunately none of these channels have been lost.**



Results of Test 2

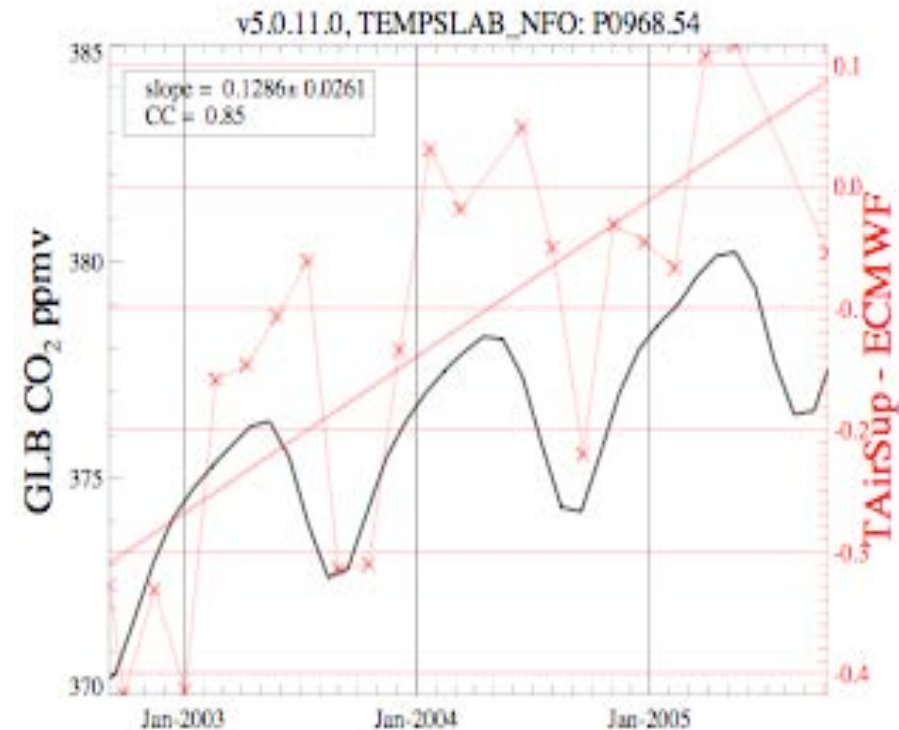
- **Differences caused by filling could be interpreted as climate trends**

Field	Change	Spurious Trend / 4 yrs
TSurfStd	+0.010 K	2.5 mK/yr warming
TSurfAir	-0.012 K	3 mK/yr cooling
TSurf1Ret	+0.026 K	6 mK/yr warming
O ₃ yield	0.4% decrease	0.1 % /yr decrease
Initial_CC_score	0.0016	0.0004 /yr increase
nchan_big_ang_adj	-0.02	0.5%/yr decrease

- **Channel filling is not the main source of the spurious trends identified by Divakarla & Hearty (~100 mK/yr)**
- **But channel filling error is significant at the level of climate: ~10 mK/yr**
- **Effect on outliers not yet evaluated**



- CO₂ changes with seasonal cycle and secular trend
- AIRS Level-2 trends resemble change in CO₂
- R-Branch CO₂ trend equivalent to ~50 mK/yr (per HHA)
- V5 regressions were trained on early mission data when CO₂ was lower
- V6 regressions will be trained to compensate



Hearty



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Plans for v6

- **Limit use of regression**
- **Train regressions to handle the entire mission**
- **Improve channel filling**



V6 Plans -- Limit Use of Regressions

- **Angle Adjustment**
 - Evaluate simple training mean adjustment, segmented by angle and perhaps day/night, latitude, land/sea, solar zenith angle, etc. but not radiances
 - Joel Susskind is exploring 3x1 retrieval with no local angle adjustment
- **Cloudy regression, clear regression, surface regression**
 - Evaluate complete removal
 - MW first guess instead of cloudy regression
 - Mini-physical retrieval instead of clear regression
 - Surface emissivity guess from MODIS historical or climatology + MW for snow detection
 - Remove PC scores and differences of regression results from other retrieved states from error estimation and quality control



V6 Plans -- Train Regressions to Handle Entire Mission

- **Representative training sets to be selected by AIRS project and NOAA science team member**
 - Will be isolated from test data
- **Revisit channel selection to use only channels sensitive to target species**
- **May need multiple epochs to cover the entire mission**
 - This increases effort
 - Smooth transitions to avoid step functions at epoch boundaries
 - Must be careful of changes in models
- **Evaluate making regressions aware of CO₂**
 - Use time as a predictor
 - Use modeled CO₂ as a predictor

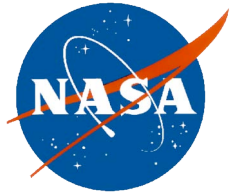


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V6 Plans -- Improve Channel Filling

- **Will not fill as many channels**
 - Use values when noise level increases but is still under ~ 1 K
 - Eliminate lower limit of 150 K on cloud cleared radiances
- **Improve channel filling algorithms**
 - Evaluate alternate algorithms
 - First guess:
 - From all nearby channels, not just those used in regression
 - From channels selected for high correlation
 - From computed radiances based on the current state
 - Multiple passes through PC



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Backup Slides

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Cloudy and Clear Profile Regressions, Surface Regression

- **Cloudy regression was added in v5 as a partial replacement for MW-only retrieval**
- **It is used as a first guess of profile into the first iteration of cloud clearing**
- **Clear profile regression runs after first cloud clearing**
 - It provides a profile for use in the second iteration of cloud clearing.
 - Its fine vertical structure is preserved in physical retrieval retrieval
- **Surface regression runs right after clear profile regression.**
 - It provides an estimate of surface spectral emissivity used in second cloud clearing
 - Its fine structure is preserved through physical retrieval



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Why Channels Are Filled

- **Noise levels of individual channels can change**
- **Some detectors have experienced significant long-term changes in noise levels**
 - See presentation by Denis Elliott
- **Other channels experience occasional changes in dark current ("pops") or transient high-noise events**
- **2-point calibration prevents any changes in bias -- only noise level changes**



Why Channels Are Filled

- **Training a regression implicitly makes it expect a given noise level -- it weights lower-noise channels more heavily**
- **Channels experiencing significantly higher noise levels than they had in the training set are not used as input to regressions**
- **But the regressions need input for all channels.**
- **Channel filling algorithms replace the radiance of a missing channel with a predicted radiance**
- **But note: the current screening of channels appears to be too strict, leading to too much channel filling. Version 6 will depend less on this method.**



- **Different channel filling algorithms are used in different regressions.**
- **Local angle correction**
 - Initial radiance for filled channels set to training mean
 - PC scores calculated from radiances (including filled)
 - New radiances calculated from PC scores
- **Clear and cloudy profile regressions + surface regression**
 - Initial radiance for filled channels set to match mean of differences from training mean of radiances of 10 spectrally close* channels used in regression
 - PC scores calculated from radiances (including filled)
 - New radiances calculated from PC scores
- *** Channels selected may not be truly spectrally close because of gaps in the spectral coverage**

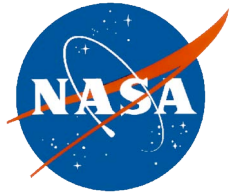


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Pitfalls of Channel Filling

- **Filled values will not have correct noise characteristics**
- **Filled values will tend toward a training mean**
 - Output will tend to be correct in an average sense but extreme cases will be curtailed
- **Because the number of channels filled tends to increase with time, results will tend to systematically exclude extreme cases with time**



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Examination of PC Score

Sung-Yung Lee

**California Institute of Technology
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Principal Component Score

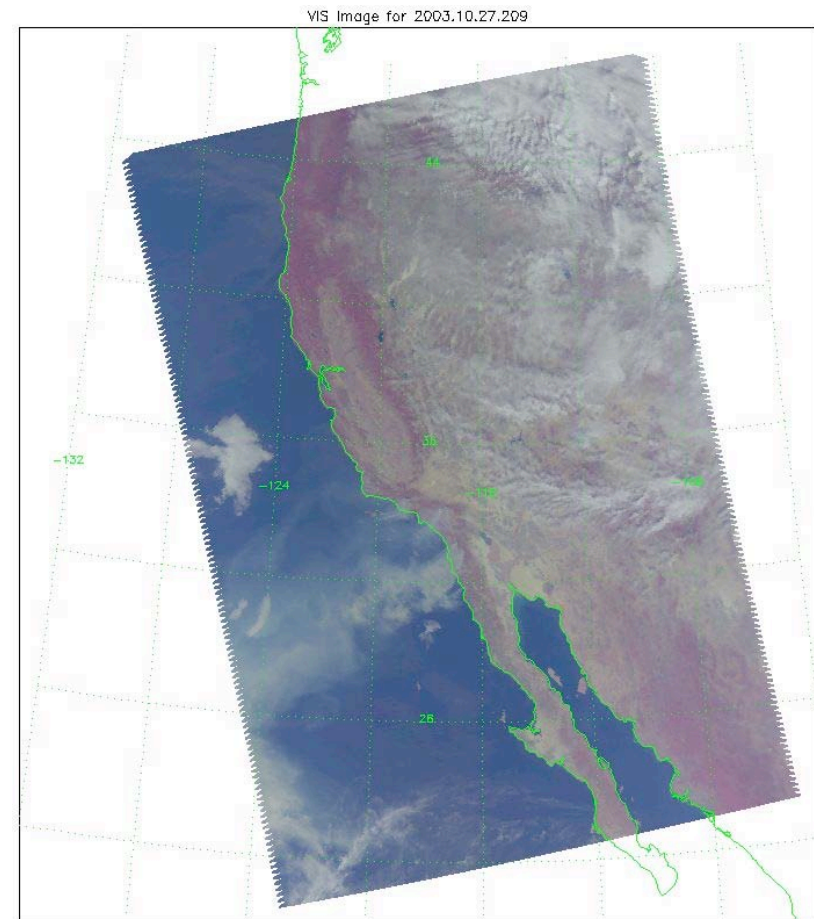
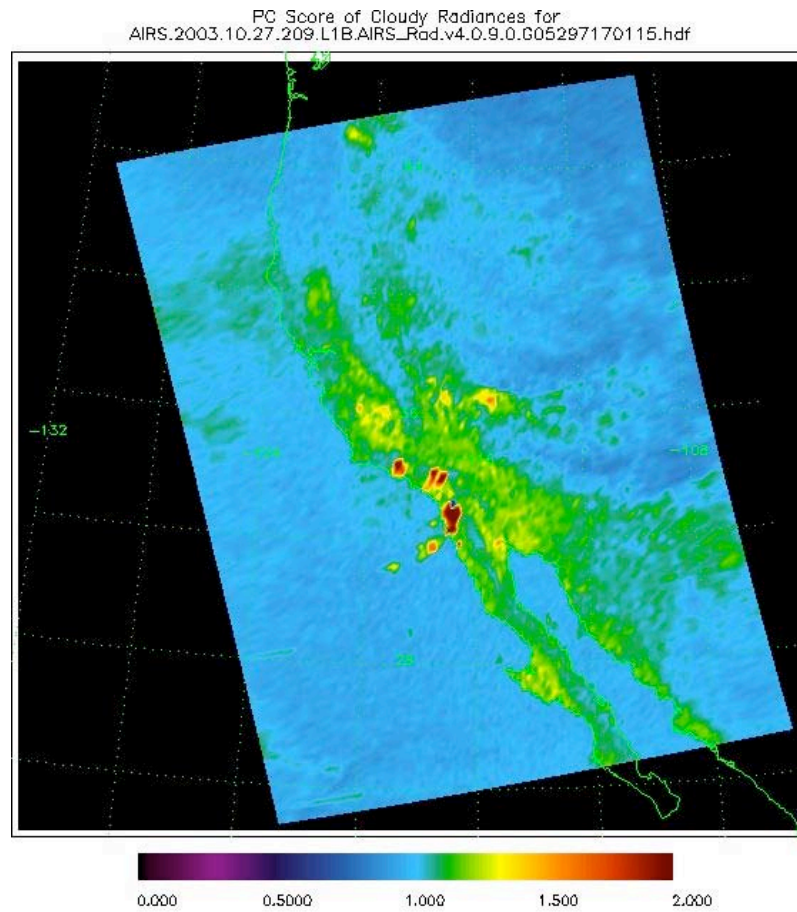
- **M Goldberg (NOAA/NESDIS) developed algorithm to compress AIRS radiances as coefficients to principal components or eigenvectors.**
 - The principal components are computed from the radiances normalized by the channel noise, NeN.
 - Many of the channels used in the PC analysis became noisy over time
 - 25 channels as of mid-2007.
 - Early report claims the channel filling algorithm is reliable with fewer than 20 bad channels.
- **The Principal component scores is defined to be the residual error of the reconstructed radiances, in the unit of NeN.**
 - PC score of AIRS observed radiance is normally around 1.
 - PC scores are large when C_{ij} becomes an issue
 - PC scores are large over sun glint area and over brush fire
 - PC scores are large over “some” dust, but not all.
- **The initial regression algorithm of AIRS uses the principal components.**
 - It is applied to cloud cleared radiances.
 - The PC score is used as a measure of quality of cloud clearing.
 - Currently retrievals are rejected when PC score is larger than 4.



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High PC Scores over Brush Fire



- Granule 209 of Oct 27, 2003

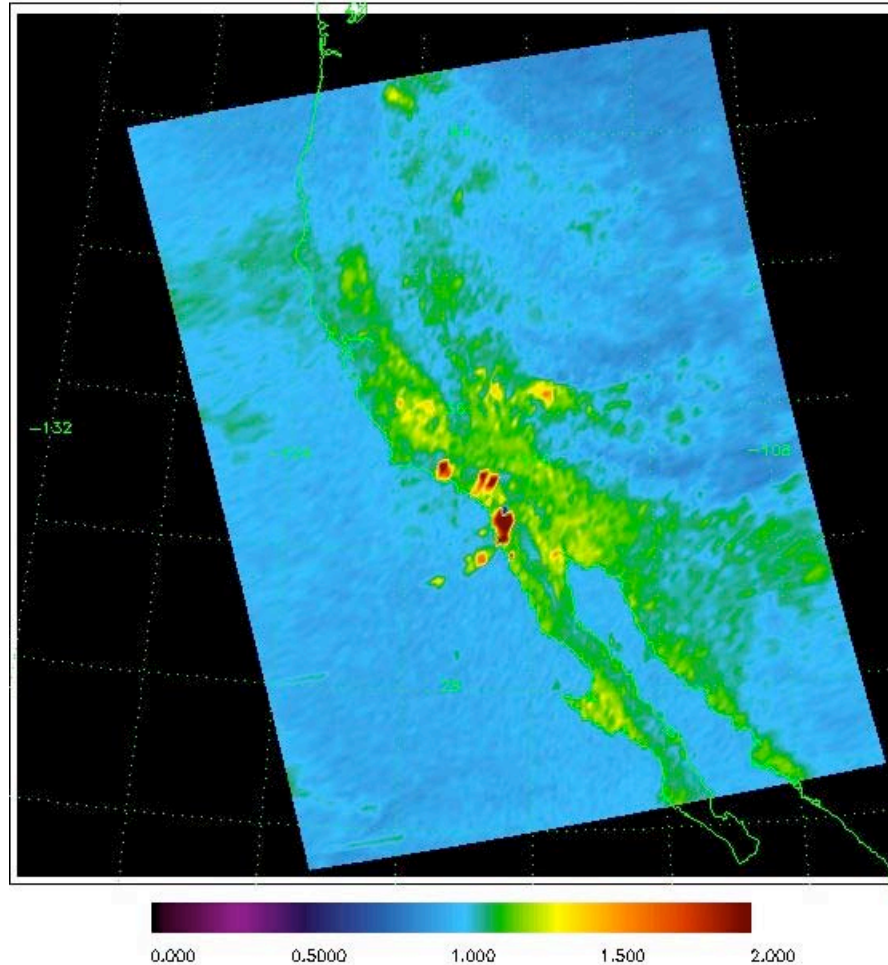


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Granule 209 of Oct 27, 2003

PC Score of Cloudy Radiances for
AIRS.2003.10.27.209.L1B.AIRS_Rad.v4.0.9.0.G05297170115.hdf



- There are many brushfires burning in Simi Valley, Mt Baldy, Arrowhead, Cleveland Forest and San Diego.
- CO plume over ocean also has high values of PC score

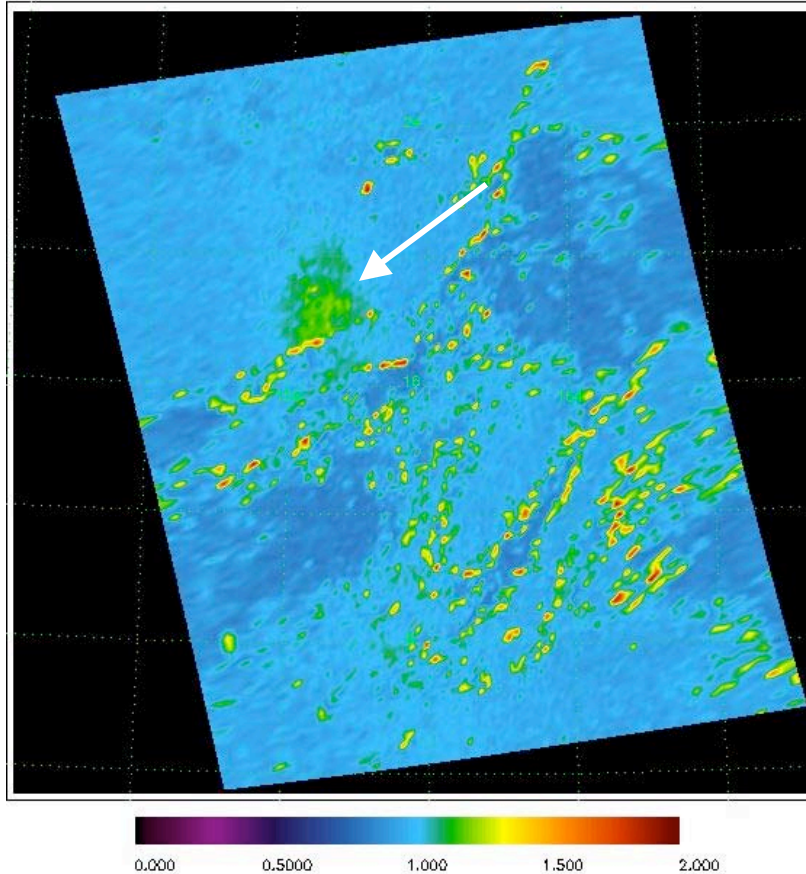


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Granule 27 of Sept 6, 2007

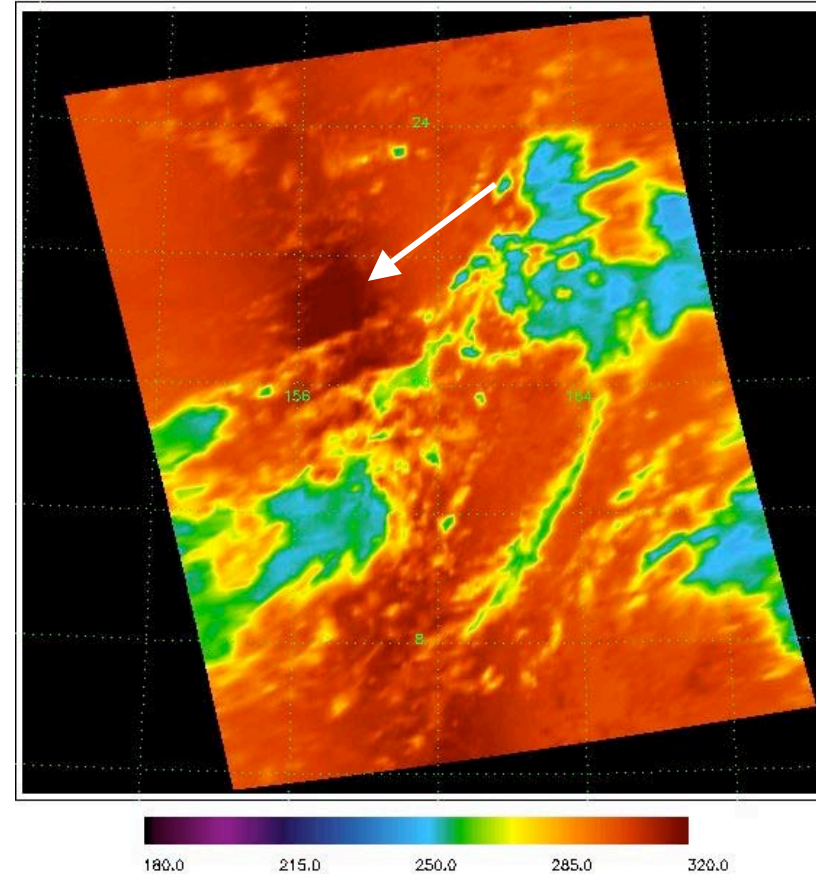
PC Score of Cloudy Radiances for
AIRS.2002.09.06.027.L1B.AIRS_Rad.v4.8.1.0.Focus.T06298153527.hdf



PC Score

- Clear sun glint area has higher value of PC score

Brightness Temperature of 2616.4 cm^{-1} Channel for
AIRS.2002.09.06.027.L1B.AIRS_Rad.v4.8.1.0.Focus.T06298153527.hdf



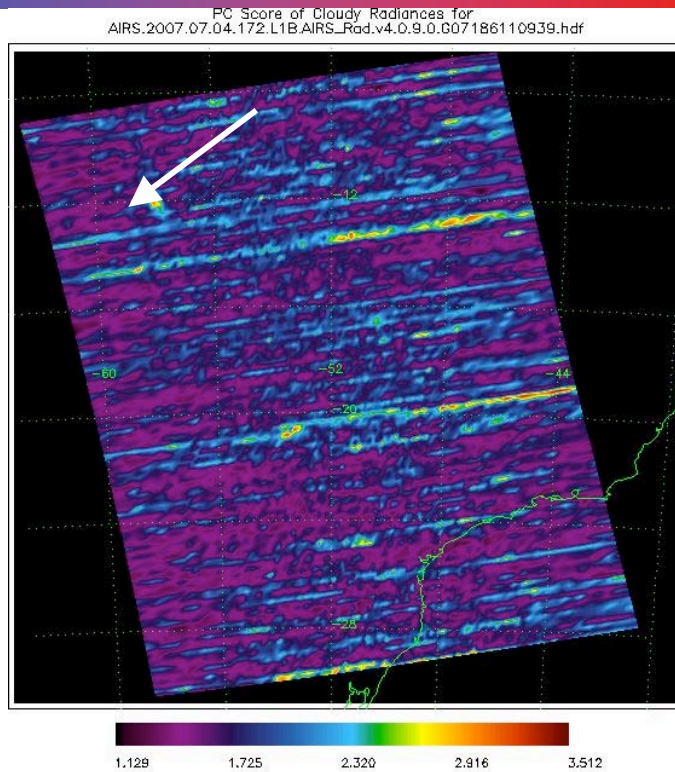
Tb 2616 cm^{-1}



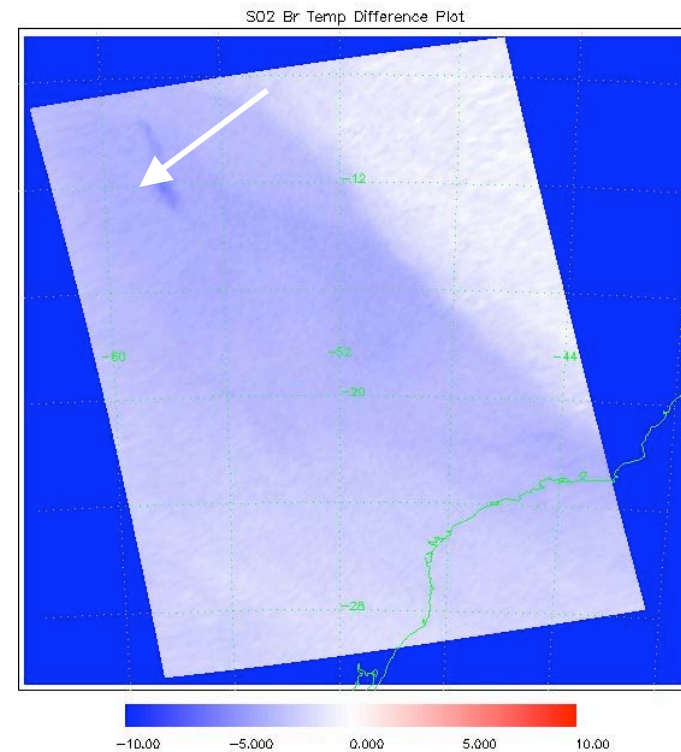
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Granule 172 of July 4, 2007



PC Score

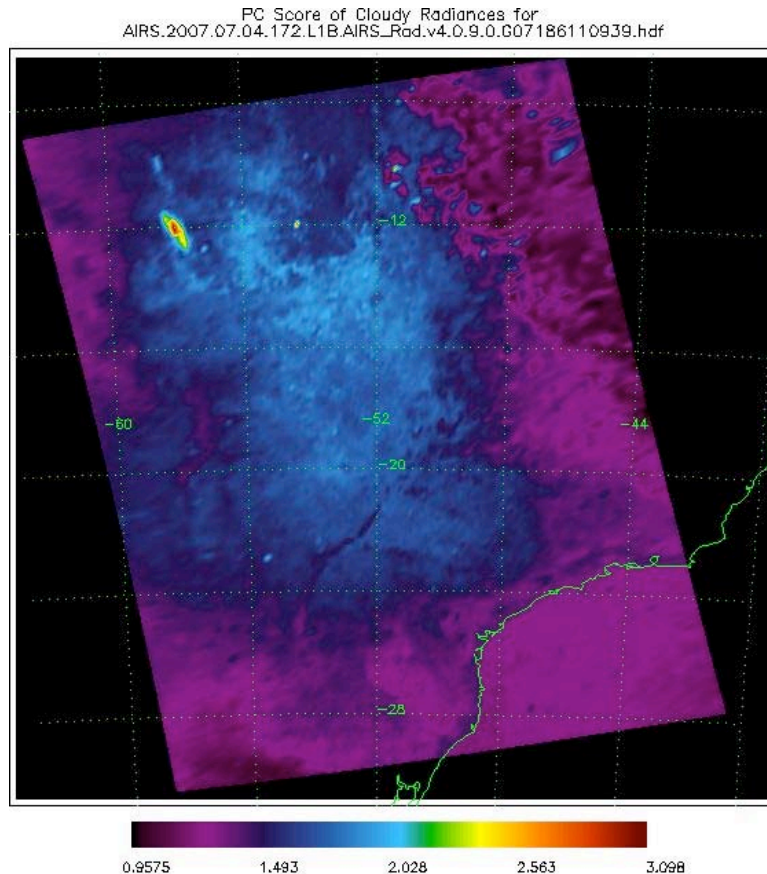


SO₂ Br Temp Diff

- PC score (this version does not fill bad channels) degraded due to channel losses over time (29 bad channels in this granule)
- SO₂ plume (yet unknown source) is noticeable even in this figure



Granule 172 of July 4, 2007 (Continued)



- **Bad channels are filled using algorithm developed for local angle correction**
 - Fill bad channels with training mean
 - Reconstruct radiances using PC analysis
 - Fill bad channels with reconstructed radiance
 - Do another PC analysis to reconstruct
 - Compute PC score based only on good channels
- **The filled PC score clearly shows SO₂ plume**



Contingency -- Use Level-1C

- **Level-1C product will contain radiances resampled to a fixed grid.**
 - This eliminates a minor issue with regressions
- **Level-1C may also include filled values**
 - This would eliminate the need for channel filling in Level-2
 - Level-1C now would be responsible for avoiding the pitfalls of channel filling
- **Following this path for version 6 brings scheduling complications:**
 - Define Level-1C algorithm
 - Implement Level-1C
 - Run Level-1C on large dataset
 - Test Level-1C
 - If Level-1C acceptable, retrain all Level-2 regressions
 - If Level-1C is not acceptable, do something else in Level-2
 - Test Level-2